

### Two Types of Discrimination

- ▶ Taste-based discrimination
  - ▶ A pure disutility for hiring, working with, or being around a certain group
  - ▶ No economic motive
  - ▶ Example?
- ▶ Statistical discrimination
  - ▶ Membership in a certain group can be correlated with other characteristics that are relevant for hiring, eg education level
  - ▶ Given this correlation, it may make sense for hiring manager to use group membership as a criteria
  - ▶ Purely economic motivation, no actual animus towards group
  - ▶ Example?

### Discrimination

### Are Emily and Greg More Employable Than Lakisha and Jamal?

- ▶ Want to examine racial discrimination in job hiring practices
- ▶ Normally race and job-relevant characteristics (education, skills, etc) may be correlated
- ▶ Need an experimental design where race is truly randomly assigned
- ▶ Research design by Bertrand and Mullainathan (2004):
  - ▶ Create many composite resumes based on real ones
  - ▶ Some are high skill, some are low skill
  - ▶ Randomly put either white-sounding or African-American-sounding name on top of each resume
  - ▶ Send resumes to real hiring managers in response to 1300 real ads
  - ▶ Send 4 resumes (1 of each type) to each
  - ▶ Measure percentage of callbacks each resume gets

## Names Used Were Distinctly Black or White

TABLE A1—FIRST NAMES USED IN EXPERIMENT

White female			African-American female		
Name	L(W)/L(B)	Perception White	Name	L(B)/L(W)	Perception Black
Allison	∞	0.926	Aisha	209	0.97
Anne	∞	0.962	Ebony	∞	0.9
Carrie	∞	0.923	Keisha	116	0.93
Emily	∞	0.925	Kenya	∞	0.967
Jill	∞	0.889	Lakisha	∞	0.967
Laurie	∞	0.963	Latonya	∞	1
Kristen	∞	0.963	Latoya	∞	1
Meredith	∞	0.926	Tamika	284	1
Sarah	∞	0.852	Tanisha	∞	1
Fraction of all births:			Fraction of all births:		
3.8 percent			7.1 percent		

White male			African-American male		
Name	L(W)/L(B)	Perception White	Name	L(B)/L(W)	Perception Black
Brad	∞	1	Darnell	∞	0.967
Brendan	∞	0.667	Hakim	∞	0.933
Geoffrey	∞	0.731	Jamal	257	0.967
Greg	∞	1	Jermaine	90.5	1
Brett	∞	0.923	Kareem	∞	0.967
Jay	∞	0.926	Leroy	44.5	0.933
Matthew	∞	0.888	Rashheed	∞	0.931
Neil	∞	0.654	Tremayne	∞	0.897
Todd	∞	0.926	Tyrone	62.5	0.900
Fraction of all births:			Fraction of all births:		
1.7 percent			3.1 percent		

## Evidence for Discrimination

TABLE 1—MEAN CALLBACK RATES BY RACIAL SOUNDINGNESS OF NAMES

	Percent callback for White names	Percent callback for African-American names	Ratio	Percent difference ( <i>p</i> -value)
Sample:				
All sent resumes	9.65	6.45	1.50	3.20
	[2,435]	[2,435]		(0.0000)
Chicago	8.06	5.40	1.49	2.66
	[1,352]	[1,352]		(0.0057)
Boston	11.63	7.76	1.50	4.05
	[1,083]	[1,083]		(0.0023)
Females	9.89	6.63	1.49	3.26
	[1,860]	[1,886]		(0.0003)
Females in administrative jobs	10.46	6.55	1.60	3.91
	[1,358]	[1,359]		(0.0003)
Females in sales jobs	8.37	6.83	1.22	1.54
	[502]	[527]		(0.3523)
Males	8.87	5.83	1.52	3.04
	[575]	[549]		(0.0513)

### ► Summary?

5 / 19

6 / 19

## Effect of Resume Characteristics

TABLE 5—EFFECT OF RESUME CHARACTERISTICS ON LIKELIHOOD OF CALLBACK

Dependent Variable: Callback Dummy			
Sample:	All resumes	White names	African-American names
Years of experience (*10)	0.07 (0.03)	0.13 (0.04)	0.02 (0.03)
Years of experience? (*100)	-0.02 (0.01)	-0.04 (0.01)	-0.00 (0.01)
Volunteering? (Y = 1)	-0.01 (0.01)	-0.01 (0.01)	0.01 (0.01)
Military experience? (Y = 1)	-0.00 (0.01)	0.02 (0.03)	-0.01 (0.02)
E-mail? (Y = 1)	0.02 (0.01)	0.03 (0.01)	-0.00 (0.01)
Employment holes? (Y = 1)	0.02 (0.01)	0.03 (0.02)	0.01 (0.01)
Work in school? (Y = 1)	0.01 (0.01)	0.02 (0.01)	-0.00 (0.01)
Honors? (Y = 1)	0.05 (0.02)	0.06 (0.03)	0.03 (0.02)
Computer skills? (Y = 1)	-0.02 (0.01)	-0.04 (0.02)	-0.00 (0.01)
Special skills? (Y = 1)	0.05 (0.01)	0.06 (0.02)	0.04 (0.01)
<i>H<sub>0</sub></i> : Resume characteristics effects are all zero ( <i>p</i> -value)	54.50 (0.0000)	57.59 (0.0000)	23.85 (0.0080)
Standard deviation of predicted callback	0.047	0.062	0.037
Sample size	4,870	2,435	2,435

### ► Summary?

## Alternate Explanations for Results

### ► What are other explanations for apparent taste-based racial discrimination?

7 / 19

8 / 19

## Motivation

### Gender

- ▶ So far we have focused in this class mostly on behavior of an entire population
- ▶ However, lots of evidence in economics of *individual differences* in race, gender, age, etc
- ▶ Gender is correlated with different risk preferences and social preferences, for example
- ▶ Gender especially easy to study because it is randomly assigned

9 / 19

10 / 19

## Risk Aversion Differences Between Men and Women

- ▶ Experiment by Eckel and Grossman (2002)
- ▶ Subjects choose one of five risky options
  - ▶ Choice 1 is lowest risk and lowest expected payoff
  - ▶ Choice 5 is highest risk and highest expected payoff
- ▶ Two framings
  - ▶ Loss frame: paid \$6 for completing experiment
  - ▶ Gain frame: no fixed payment

Table 1  
Gamble choices, expected payoffs, and risk in the two alternative framings

Gamble choice	Event	Probability (%)	Payoff		Expected payoff		Risk
			Loss framing (\$)	No-Loss framing (\$)	Loss framing (\$)	No-Loss framing (\$)	
1	A	50	10	16	10	16	0.00
	B	50	10	16			
2	A	50	18	24	12	18	4.24
	B	50	6	12			
3	A	50	26	32	14	20	8.48
	B	50	2	8			
4	A	50	34	40	16	22	12.73
	B	50	-2	4			
5	A	50	42	48	18	24	16.97
	B	50	-6	0			

The level of risk is represented as the S.D. of expected payoff.

## Men's Choices vs Women's Choices

Table 2  
Frequency distributions of gamble choices in relation to the subject's sex and the framing treatment

Gamble choice	All subjects		Men		Women	
	Loss framing	No-Loss framing	Loss framing	No-Loss framing	Loss framing	No-Loss framing
1	7	3	2	0	5	3
2	25	10	11	6	14	4
3	48	17	15	10	33	7
4	32	9	18	6	14	3
5	36	13	26	10	10	3
Total	148	52	72	32	76	20
Mean gamble choice (S.D.)	3.44 (1.17)	3.37 (1.22)	3.76 (1.18)	3.63 (1.13)	3.14 (1.08)	2.95 (1.28)

- ▶ Summary of these results?
- ▶ Question: can we say this is due entirely to biology?

11 / 19

12 / 19

## More Motivation

- ▶ We see employment differences between men and women in many dimensions
  - ▶ Wages
  - ▶ Choice of job
  - ▶ Choice to work at all
- ▶ What causes these differences?

13 / 19

## Gender Differences in Competition

- ▶ Research design by Gneezy, Niederle, and Rustichini (2003)
  - ▶ Undergraduate engineering students
  - ▶ Groups of 6 students (3 men, 3 women)
  - ▶ Task: solving mazes of varying difficulty on the computer
- ▶ Two treatments (piece rate):
  1. Non-competitive
    - ▶ Paid 2 dollars for every solved maze
    - ▶ Score is private
  2. Competitive (tournament):
    - ▶ Person that solves most mazes gets 12 dollars for each maze solved
    - ▶ All others in group receive nothing
    - ▶ Winner anonymous

14 / 19

## Performance by Gender in Piece Rate

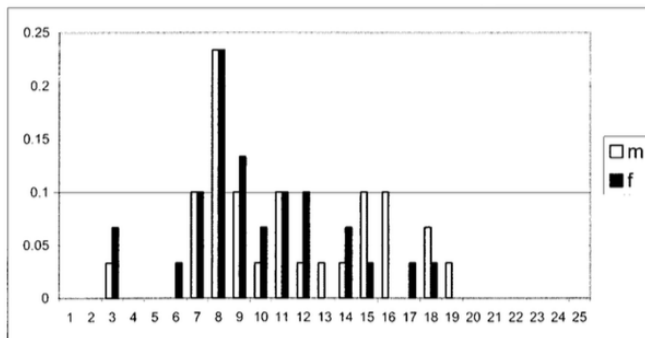


FIGURE I  
Number of Mazes Solved under Piece Rate

15 / 19

## Performance by Gender in Tournament

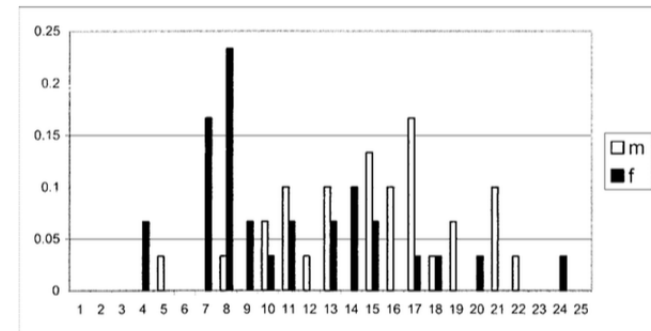


FIGURE II  
Number of Mazes Solved under Tournament Condition

16 / 19

## Gender Gap

- ▶ In summary:
  - ▶ Small, statistically insignificant gender gap under piece rate (11.23 vs 9.73,  $p = 0.202$ )
  - ▶ Larger, statistically significant gender gap under tournament (15.00 vs 10.9,  $p < .01$ )
- ▶ What could be causing this performance gender gap in one setting but not the other?

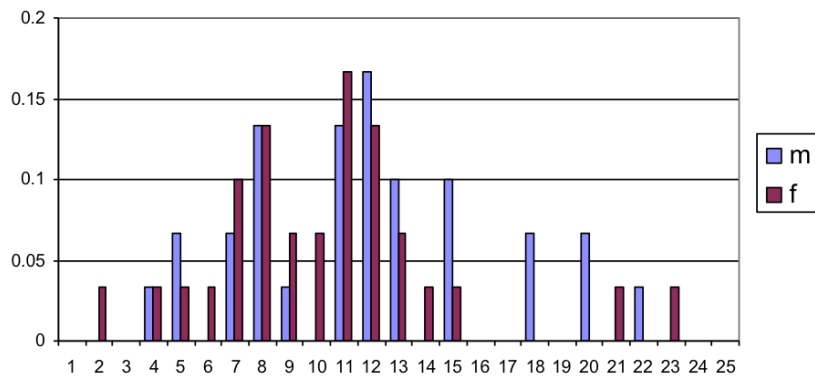
17 / 19

## Two Additional Treatments

1. Uncertain payment
  - ▶ One person chosen at random and paid 12 dollars for each maze solved correctly
  - ▶ Score is private
2. Single-sex tournament:
  - ▶ Groups of all 6 men or all 6 women
  - ▶ Person that solves most mazes gets 12 dollars for each maze solved
  - ▶ All others in group receive nothing
  - ▶ Winner anonymous

18 / 19

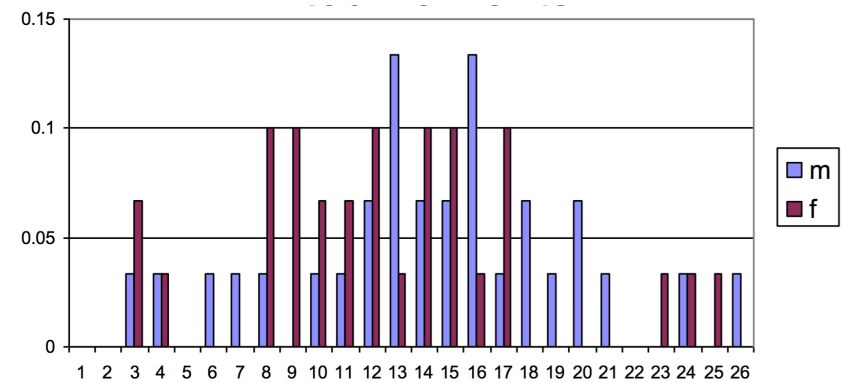
## Uncertain Payment



Mean for males: 11.83, for females: 10.33.  $p = 0.165$

19 / 19

## Single-Sex Tournaments



Men: 14.3, Women: 12.6,  $p = 0.135$

20 / 19

## Summary of Results

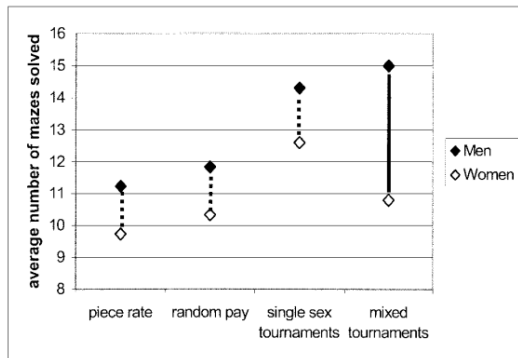


FIGURE III

Averages Performance of the 30 Men and 30 Women in Each of the Treatments

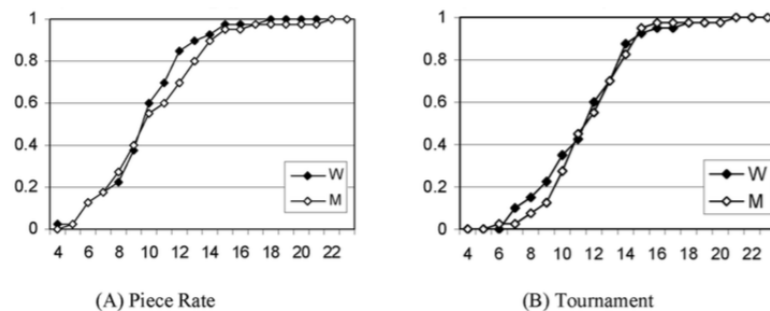
21 / 19

## Selection into Competitive Environments

- ▶ Main results from previous paper: significant gender gap seems to exist only when women are competing directly against men
- ▶ Natural question: are women aware of this preference, and do they consider it when choosing which environments to enter?
- ▶ Research design by Niederle and Vesterlund (2007):
  - ▶ Groups of 4 (2 men, 2 women)
  - ▶ Different task: add groups of 5 two-digit numbers
  - ▶ As before, two treatments: piece-rate (50 cents per correct answer) and tournament (2 dollars per correct answer for winner only)
  - ▶ Initially, subjects randomly assigned into a treatment

22 / 19

## Baseline Results: No Gender Gap in Performance



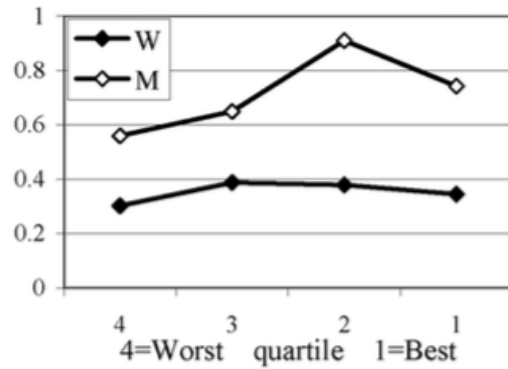
23 / 19

## Selection Into Tournament

- ▶ After 5 rounds of either piece-rate or tournament, subjects get to choose between the two for the next part of the study
- ▶ Based on performance we see in baseline, women and men are expected to do equally well in the tournament
  - ▶ Top 30% of both genders should choose tournament
- ▶ What actually happens?
  - % of women choose tournament
  - % of men choose tournament

24 / 19

## Likelihood to Enter Tournament vs Ranking



► Summary?

## What Could Cause Difference?

- What could be causing the difference in entrance rates?
- Perhaps women have lower confidence in their own rank
- So, authors ask subjects to report what they think their rank is within their group of 4
  - Paid 1 dollar if correct, nothing otherwise

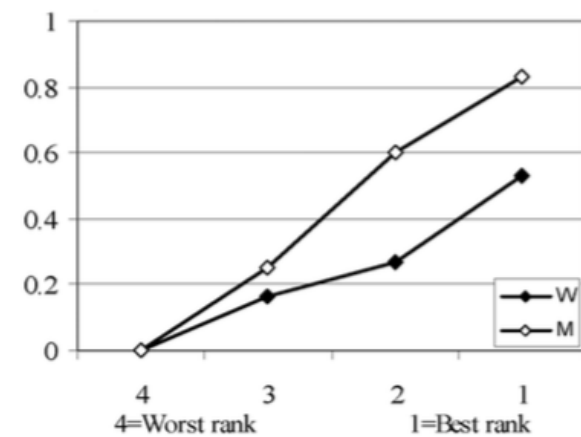
25 / 19

26 / 19

## Men Supremely Over-Confident

DISTRIBUTION OF GUESSED TOURNAMENT RANK				
	Men		Women	
	Guessed rank	Incorrect guess	Guessed rank	Incorrect guess
1: Best	30	22	17	9
2	5	3	15	10
3	4	2	6	5
4: Worst	1	1	2	1
Total	40	28	40	25

## Likelihood to Enter Tournament vs *Guessed* Ranking



► Summary?

27 / 19

28 / 19